

# COURSE 3

## 2.2. Lagrange interpolation (continuation)

Let  $[a, b] \subset \mathbb{R}$ ,  $x_i \in [a, b]$ ,  $i = 0, 1, \dots, m$  such that  $x_i \neq x_j$  for  $i \neq j$  and consider  $f : [a, b] \rightarrow \mathbb{R}$ .

The Lagrange polynomial generates **the Lagrange interpolation formula**

$$f = L_m f + R_m f,$$

where  $R_m f$  denotes **the remainder (the error)**.

**Theorem 1** Let  $\alpha = \min\{x, x_0, \dots, x_m\}$  and  $\beta = \max\{x, x_0, \dots, x_m\}$ . If  $f \in C^m[\alpha, \beta]$  and  $f^{(m)}$  is derivable on  $(\alpha, \beta)$  then  $\forall x \in (\alpha, \beta)$ , there exists  $\xi \in (\alpha, \beta)$  such that

$$(R_m f)(x) = \frac{u(x)}{(m+1)!} f^{(m+1)}(\xi). \quad (1)$$

**Corollary 2** If  $f \in C^{m+1}[a, b]$  then

$$|(R_m f)(x)| \leq \frac{|u(x)|}{(m+1)!} \|f^{(m+1)}\|_{\infty}, \quad x \in [a, b]$$

where  $\|\cdot\|_{\infty}$  denotes the uniform norm, and  $\|f\|_{\infty} = \max_{x \in [a, b]} |f(x)|$ .

**Example 3** Which is the limit of the error for computing  $\sqrt{115}$  using Lagrange interpolation formula for the nodes  $x_0 = 100$ ,  $x_1 = 121$  and  $x_2 = 144$ ? Find the approximative value of  $\sqrt{115}$ .

**Example 4** If we know that  $\lg 2 = 0.301$ ,  $\lg 3 = 0.477$ ,  $\lg 5 = 0.699$ , find  $\lg 76$ . Study the approximation error.



## The Aitken's algorithm

Let  $[a, b] \subset \mathbb{R}$ ,  $x_i \in [a, b]$ ,  $i = 0, 1, \dots, m$  such that  $x_i \neq x_j$  for  $i \neq j$  and consider  $f : [a, b] \rightarrow \mathbb{R}$ .

Usually, for a practical approximation problem, for a given function  $f : [a, b] \rightarrow \mathbb{R}$  we have to find the approximation of  $f(\alpha)$ ,  $\alpha \in [a, b]$  with an error not greater than a given  $\varepsilon > 0$ .

If we have enough information about  $f$  and its derivatives, we use the inequality  $|R_m f(x)| \leq \varepsilon$  to find  $m$  such that  $(L_m f)(\alpha)$  approximates  $f(\alpha)$  with the given precision.

We may use the condition  $\frac{|u(x)|}{(m+1)!} \|f^{(m+1)}\|_{\infty} \leq \varepsilon$ , but it should be known  $\|f^{(m+1)}\|_{\infty}$  or a majorant of it.

A practical method for computing the Lagrange polynomial is **the Aitken's algorithm**. This consists in generating the table:

$x_0$	$f_{00}$						
$x_1$	$f_{10}$	$f_{11}$					
$x_2$	$f_{20}$	$f_{21}$	$f_{22}$				
$x_3$	$f_{30}$	$f_{31}$	$f_{32}$	$f_{33}$			
:	:	:	:	:			
$x_m$	$f_{m0}$	$f_{m1}$	$f_{m2}$	$f_{m3}$	$\dots$	$f_{mm}$	

where

$$f_{i0} = f(x_i), \quad i = 0, 1, \dots, m,$$

and

$$f_{i,j+1} = \frac{1}{x_i - x_j} \begin{vmatrix} f_{jj} & x_j - x \\ f_{ij} & x_i - x \end{vmatrix}, \quad i = 0, 1, \dots, m; j = 0, \dots, i-1.$$

For example,

$$\begin{aligned}
 f_{11} &= \frac{1}{x_1 - x_0} \begin{vmatrix} f_{00} & x_0 - x \\ f_{10} & x_1 - x \end{vmatrix} \\
 &= \frac{1}{x_1 - x_0} [f_{00}(x_1 - x) - f_{10}(x_0 - x)] \\
 &= \frac{x - x_1}{x_0 - x_1} f(x_0) + \frac{x - x_0}{x_1 - x_0} f(x_1) = (L_1 f)(x),
 \end{aligned}$$

so  $f_{11}$  is the value in  $x$  of Lagrange polynomial for the nodes  $x_0, x_1$ . We have

$$f_{ii} = (L_i f)(x),$$

$L_i f$  being Lagrange polynomial for the nodes  $x_0, x_1, \dots, x_i$ .

So  $f_{11}, f_{22}, \dots, f_{ii}, \dots, f_{mm}$  is a sequence of approximations of  $f(x)$ .

If the interpolation procedure is convergent then the sequence is also convergent, i.e.,  $\lim_{m \rightarrow \infty} f_{mm} = f(x)$ . By Cauchy convergence criterion it follows

$$\lim_{i \rightarrow \infty} |f_{ii} - f_{i-1,i-1}| = 0.$$

This could be used as a stopping criterion, i.e.,

$$|f_{ii} - f_{i-1,i-1}| \leq \varepsilon, \quad \text{for a given precision } \varepsilon > 0.$$

Recommendation is to sort the nodes  $x_0, x_1, \dots, x_m$  with respect to the distance to  $x$ , such that

$$|x_i - x| \leq |x_j - x| \quad \text{if } i < j, \quad i, j = 1, \dots, m.$$

**Example 5** Approximate  $\sqrt{115}$  with precision  $\varepsilon = 10^{-3}$ , using Aitken's algorithm.

### Newton interpolation polynomial

A useful representation for Lagrange interpolation polynomial is

$$(L_m f)(x) := (N_m f)(x) = f(x_0) + \sum_{i=1}^m (x - x_0) \dots (x - x_{i-1}) (D^i f)(x_0) \tag{2}$$

$$= f(x_0) + \sum_{i=1}^m (x - x_0) \dots (x - x_{i-1}) [x_0, \dots, x_i; f],$$

which is called **Newton interpolation polynomial**; where  $(D^i f)(x_0)$  (or denoted  $[x_0, \dots, x_i; f]$ ) is the  $i$ -th order divided difference of the function  $f$  at  $x_0$ , given by the table

	$f$	$Df$	$D^2f$	...	$D^{m-1}f$	$D^mf$
$x_0$	$f_0$	$Df_0$	$D^2f_0$	...	$D^{m-1}f_0$	$D^mf_0$
$x_1$	$f_1$	$Df_1$	$D^2f_1$		$D^{m-1}f_1$	
$x_2$	$f_2$	$Df_2$	$D^2f_2$			
...	...	...				
$x_{m-2}$	$f_{m-2}$	$Df_{m-2}$	$D^2f_{m-2}$			
$x_{m-1}$	$f_{m-1}$	$Df_{m-1}$				
$x_m$	$f_m$					

**Newton interpolation formula** is

$$f = N_m f + R_m f,$$

where  $R_m f$  denotes the remainder.

Assume that we add the point  $(x, f(x))$  at the top of the table of divided differences:

	$f$	$Df$	...	$D^{m+1}f$
$x$	$f(x)$	$(Df)(x) = [x, x_0; f]$		$[x, x_0, \dots, x_m; f]$
$x_0$	$f(x_0)$	$(Df)(x_0) = [x_0, x_1; f]$	...	
$x_1$	$f(x_1)$	$(Df)(x_1) = [x_1, x_2; f]$		
...	...	...		
$x_{m-1}$	$f(x_{m-1})$	$(Df)(x_{m-1}) = [x_{m-1}, x_m; f]$		
$x_m$	$f(x_m)$			

For obtaining the interpolation polynomial we consider

$$[x, x_0; f] = \frac{f(x_0) - f(x)}{x_0 - x} \implies f(x) = f(x_0) + (x - x_0)[x, x_0; f] \quad (3)$$

$$\begin{aligned} [x, x_0, x_1; f] &= \frac{[x_0, x_1; f] - [x, x_0; f]}{x_1 - x} \\ &\implies [x, x_0; f] = [x_0, x_1; f] + (x - x_1)[x, x_0, x_1; f]. \end{aligned} \quad (4)$$

Inserting (4) in (3) we get

$$f(x) = f(x_0) + (x - x_0)[x_0, x_1; f] + (x - x_0)(x - x_1)[x, x_0, x_1; f].$$

If we continue eliminating the divided differences involving  $x$  in the same way, we get **Newton interpolation formula**

$$f(x) = (N_m f)(x) + (R_m f)(x),$$

with

$$(N_m f)(x) = f(x_0) + \sum_{i=1}^m (x - x_0) \dots (x - x_{i-1}) [x_0, \dots, x_i; f]$$

and the remainder (the error) given by

$$(R_m f)(x) = (x - x_0) \dots (x - x_m) [x, x_0, \dots, x_m; f]. \quad (5)$$

**Remark 6** The remainder for Lagrange interpolation formula is also given by

$$(R_m f)(x) = \frac{(x - x_0) \dots (x - x_m)}{(m+1)!} f^{(m+1)}(\xi),$$

with  $\xi$  between  $x, x_0, \dots, x_m$ , so, by (5), it follows that **the divided differences are approximations of the derivatives**

$$[x, x_0, \dots, x_m; f] = \frac{f^{(m+1)}(\xi)}{(m+1)!}.$$

**Remark 7** We notice that

$$(N_i f)(x) = (N_{i-1} f)(x) + (x - x_0) \dots (x - x_{i-1})[x_0, \dots, x_i; f]$$

so the Newton polynomials of degree 2, 3, ..., can be iteratively generated, similarly to Aitken's algorithm.

**Example 8** Find  $L_2 f$  for  $f(x) = \sin \pi x$ , and  $x_0 = 0, x_1 = \frac{1}{6}, x_2 = \frac{1}{2}$ , in both forms.

**Sol.** a) We have  $u(x) = x(x - \frac{1}{6})(x - \frac{1}{2})$ ;  $u_0(x) = (x - \frac{1}{6})(x - \frac{1}{2})$ ;  $u_1(x) = x(x - \frac{1}{2})$ ;  $u_2(x) = x(x - \frac{1}{6})$

$$\begin{aligned} (L_2 f)(x) &= \sum_{i=0}^2 l_i(x) f(x_i) = \sum_{i=0}^2 \frac{u_i(x)}{u_i(x_i)} f(x_i) \\ &= \frac{(x - \frac{1}{6})(x - \frac{1}{2})}{(-\frac{1}{6})(-\frac{1}{2})} 0 + \frac{x(x - \frac{1}{2})}{\frac{1}{6}(-\frac{1}{3})} \frac{1}{2} + \frac{x(x - \frac{1}{6})}{\frac{1}{2} \cdot \frac{1}{3}} 1 \\ &= -3x^2 + \frac{7}{2}x. \end{aligned}$$

b)

$$\begin{aligned}(N_2f)(x) &= f(0) + \sum_{i=1}^2 (x - x_0) \dots (x - x_{i-1})(D^i f)(x_0) \\&= f(0) + (x - x_0)(Df)(x_0) + (x - x_0)(x - x_1)(D^2f)(x_0) \\&= x(Df)(x_0) + x(x - \frac{1}{6})(D^2f)(x_0)\end{aligned}$$

*The table of divided differences:*

$x$	$f$	$Df$	$D^2f$
0	0	3	-3
$\frac{1}{6}$	$\frac{1}{2}$	$\frac{3}{2}$	
$\frac{1}{2}$	1		

so

$$(N_2f)(x) = 3x - 3x(x - \frac{1}{6}) = -3x^2 + \frac{7}{2}x.$$